

# Multi-Level Repetition Benchmarking

 AI CENTER

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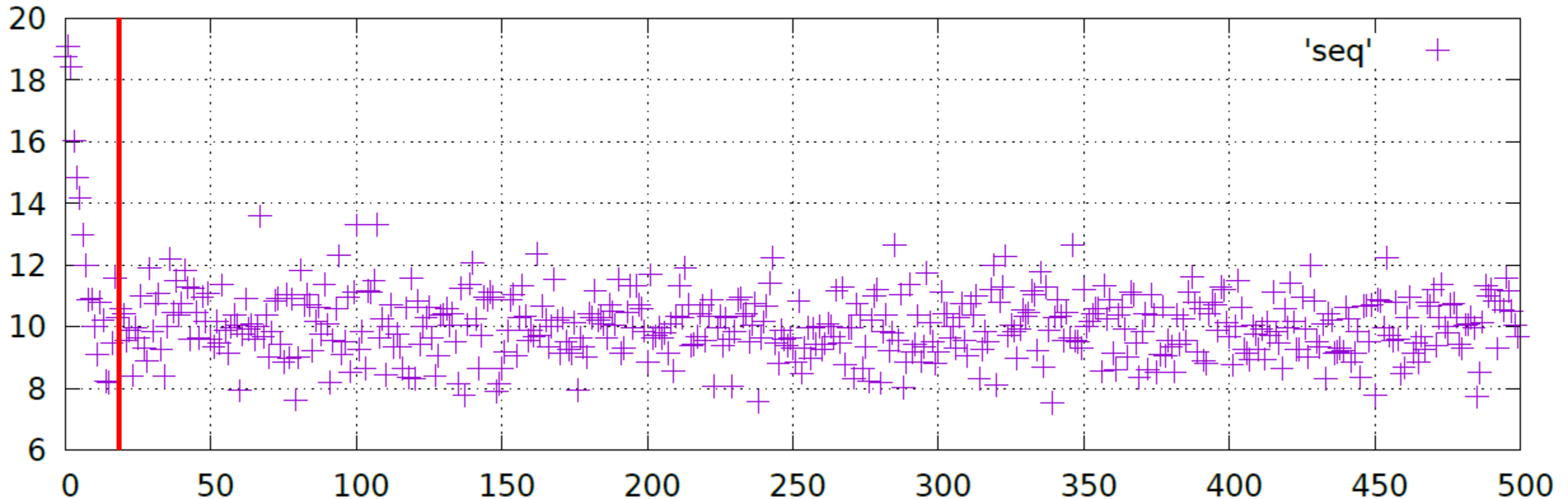


# Multi-Level Repetition

- Variance in measurements may occur at higher levels
  - we need to repeat measurements at least on the level of the variance
- The highest level is the most important one
- Levels:
  1. Iteration – loop body
  2. Execution – running of the program
  3. Compilation (stable in Java)

# Warm-up

- Measurements usefull only after reaching a steady/independent state
- E.g. by manual inspection of sequence plot



# Execution Time with Confidence Interval

$$\bar{Y} \pm t_{1-\frac{\alpha}{2}, \nu} \sqrt{\frac{1}{r_n(r_n - 1)} \sum_{j_n=1}^{r_n} \left( \bar{Y}_{j_n} \underbrace{\dots}_{n-1} - \bar{Y} \right)^2}$$

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- $(1 - \alpha)$  - confidence interval (e.g. 95% confidence  $\rightarrow \alpha = 0.05$ )
- $n$  - number of levels
- $r_n$  - number of repetition on the highest level
- $\bar{Y}$  - mean across all measurements (after reaching independent state)
- $t_{1-\frac{\alpha}{2}, \nu}$  -  $\left(1 - \frac{\alpha}{2}\right)$ -quantile of the  $t$ -distribution with  $\nu = r_n - 1$  degrees of freedom, can be found in a table

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- The mean across all measurements but the highest level for  $j_n$ -th repetition
- Example:  $n = 2, r_1 = 10, r_n = r_2 = 20$

$$\bar{Y}_5 \underbrace{\dots}_{2-1} = \frac{\sum_{i=1}^{r_1} Y_{5,i}}{r_1}$$

$Y_{a,b}$  - measured time of the  $b$ -th iteration of the  $a$ -th execution



# Speed-Up Ratios with Confidence Interval

$$\frac{\bar{Y} \cdot \bar{Y}' \mp \sqrt{(\bar{Y} \cdot \bar{Y}')^2 - (\bar{Y}^2 - h^2)(\bar{Y}'^2 - h'^2)}}{\bar{Y}^2 - h^2}$$

$$h = \sqrt{t_{\frac{\alpha}{2}, \nu}^2 \frac{S_n^2}{r_n}} \quad h' = \sqrt{t_{\frac{\alpha}{2}, \nu}^2 \frac{S_n'^2}{r_n}}$$

- $\bar{Y}, \bar{Y}'$  - means of the compared implementations
- $h, h'$  - half-widths of the confidence intervals for the single implementations
- $S_n^2, S_n'^2$  - variance estimator of the  $n$ -th level

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- $t$ -distribution is symmetric  $\rightarrow t_{\frac{\alpha}{2}, \nu}^2 = t_{1-\frac{\alpha}{2}, \nu}^2$

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# References

[1] Kalibera, T. and Jones, R. E. (2013) Rigorous Benchmarking in Reasonable Time. In: ACM SIGPLAN International Symposium on Memory Management (ISMM 2013), 20–12 June, 2013, Seattle, Washington, USA. <http://kar.kent.ac.uk/33611/>